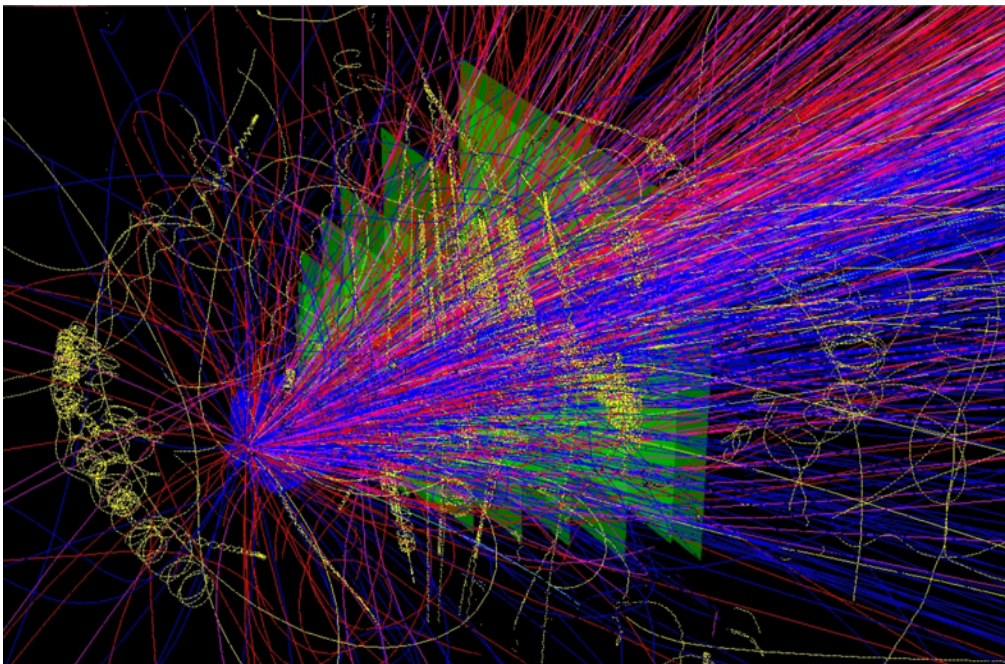


# Parallel 4-Dimensional Cellular Automaton Track Finder for the CBM Experiment

Valentina Akishina and Ivan Kisel  
for the CBM Collaboration

Goethe-University Frankfurt am Main  
FIAS Frankfurt Institute for Advanced Studies  
GSI Helmholtz Center for Heavy Ion Research

# Reconstruction Challenge in CBM at FAIR/GSI

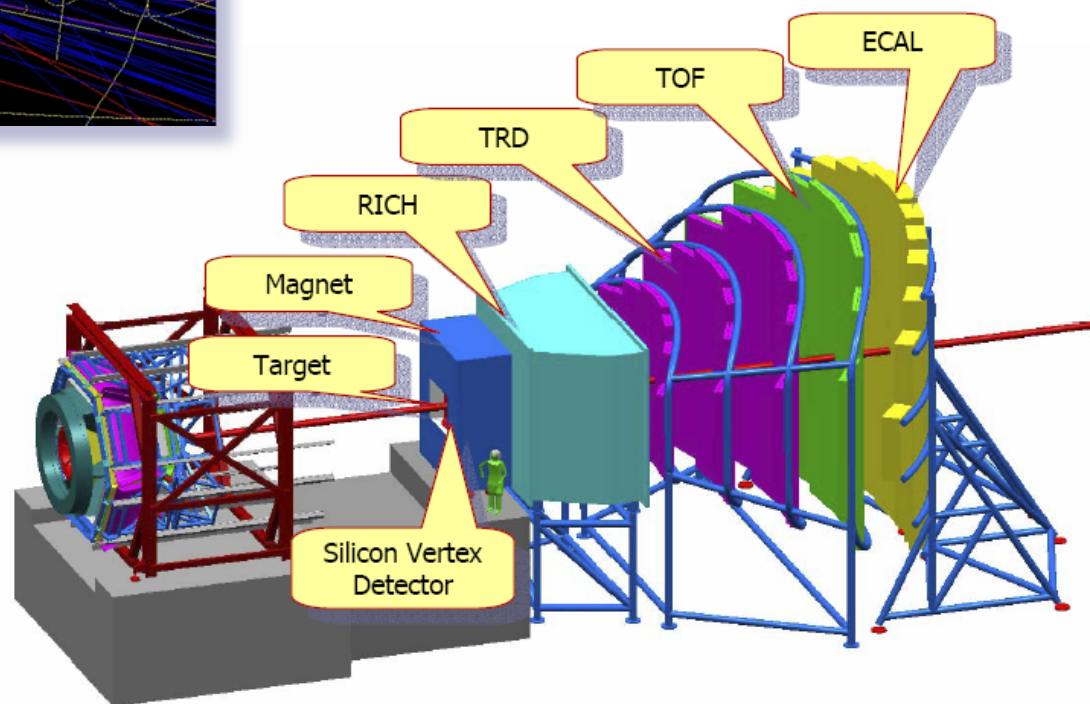


- Future **fixed-target heavy-ion** experiment
- $10^7$  **Au+Au** collisions/sec
- $\sim 1000$  charged **particles/collision**
- **Non-homogeneous** magnetic field
- **Double-sided strip detectors** (85% **fake space-points**)

Full event reconstruction will be done **on-line** at the First-Level Event Selection (**FLES**) and off-line using the same **FLES** reconstruction package.

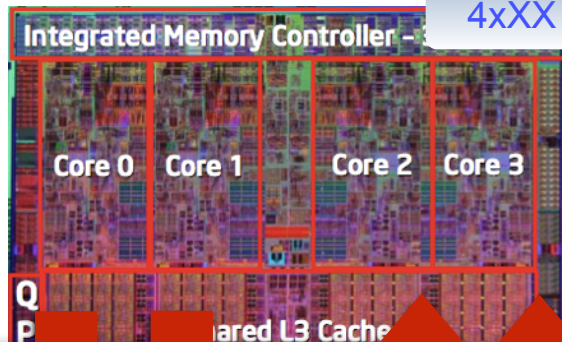
Cellular Automaton (CA) Track Finder  
Kalman Filter (KF) Track Fitter  
KF short-lived Particle Finder

All reconstruction algorithms are **vectorized** and **parallelized**.



# Many-Core CPU/GPU Architectures

Intel/AMD CPU



4xXX cores

Math

Memory

- Optimized for low latency access to cache data sets
- Control for out-of-order and speculative execution

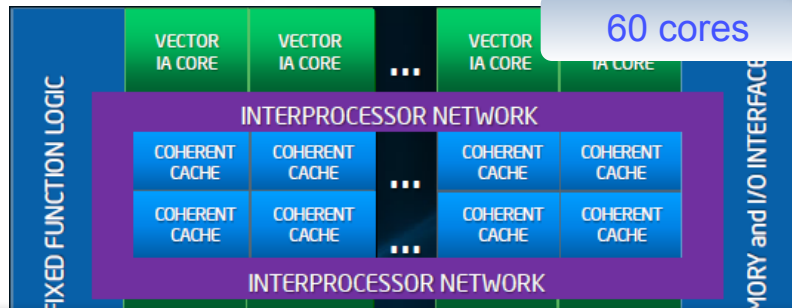
Parallelism

Math

Memory

#Cores

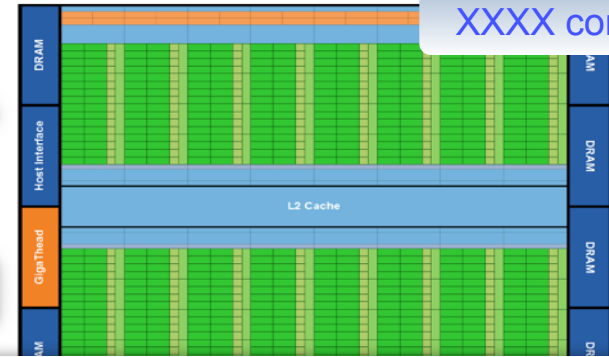
Intel Phi



60 cores

- Many Integrated Cores architecture announced at ISC10 (June 2010)
- Based on the x86 architecture
- Many-cores + 4-way multithreaded + 512-bit wide vector unit

Nvidia/ATI GPU



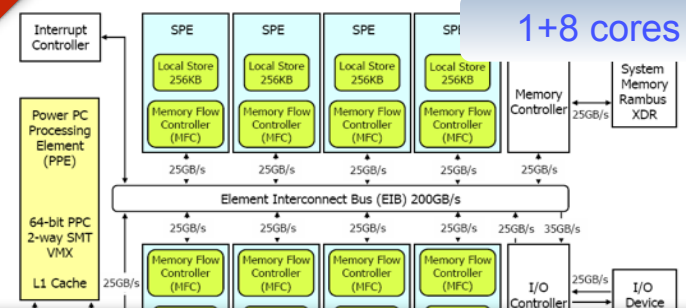
XXXX cores

- Optimized for data-parallel, throughput computation
- More transistors dedicated to computation

Stability

Memory

IBM Cell



1+8 cores

- General purpose RISC processor (PowerPC)
- 8 co-processors (SPE, Synergistic Processor Elements)
- 128-bit wide SIMD units

Future systems are heterogeneous, but using the same code

# Kalman Filter (KF) Track Fit Library

## Kalman Filter Methods

### Kalman Filter Tools:

- KF Track Fitter
- KF Track Smoother
- Deterministic Annealing Filter

### Kalman Filter Approaches:

- Conventional DP KF
- Conventional SP KF
- Square-Root SP KF
- UD-Filter SP
- Gaussian Sum Filter

### Track Propagation:

- Runge-Kutta
- Analytic Formula

## Implementations

### Vectorization (SIMD):

- Header Files
- Vc Vector Classes
- ArBB Array Building Blocks
- OpenCL

### Parallelization (many-cores):

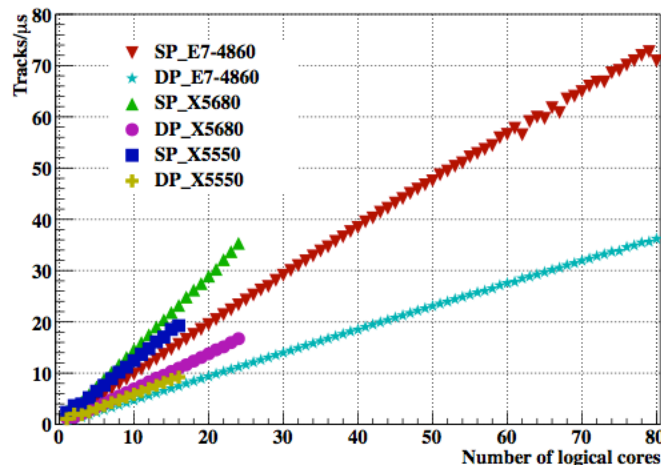
- Open MP
- ITBB
- ArBB
- OpenCL

### Precision:

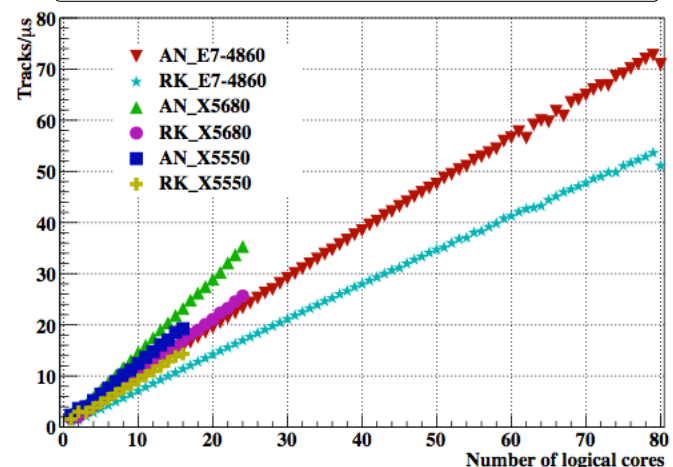
- single precision SP
- double precision DP

Comp. Phys. Comm. 178 (2008) 374-383

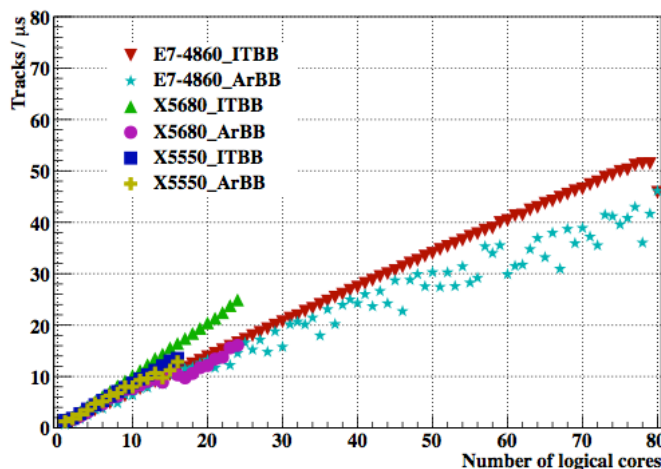
Conventional KF DP vs. SP



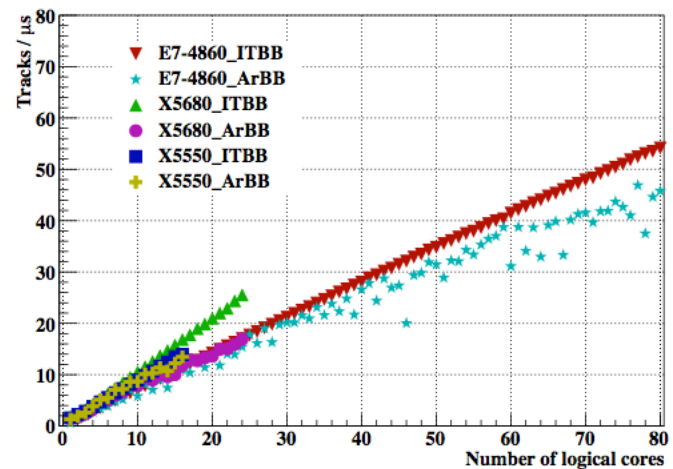
Conventional KF RK4 vs. Analytical



Square-Root KF



UD KF

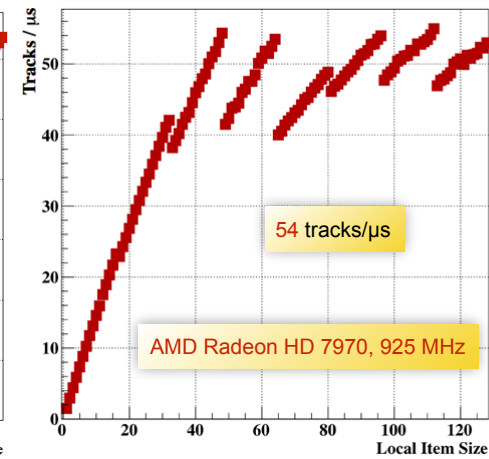
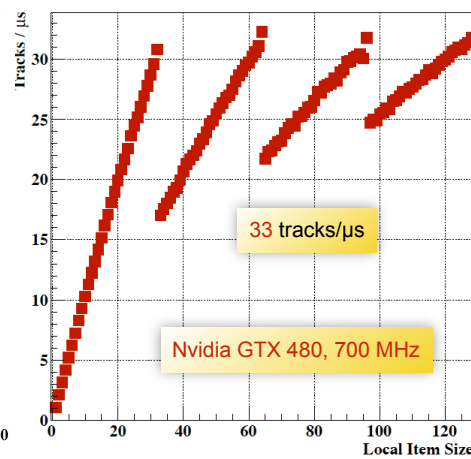
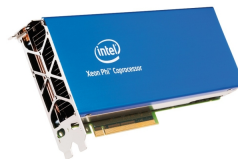
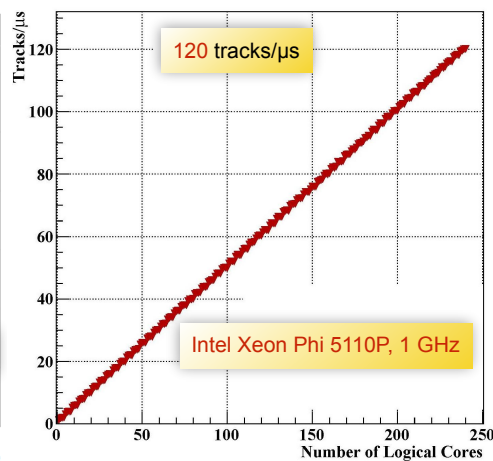
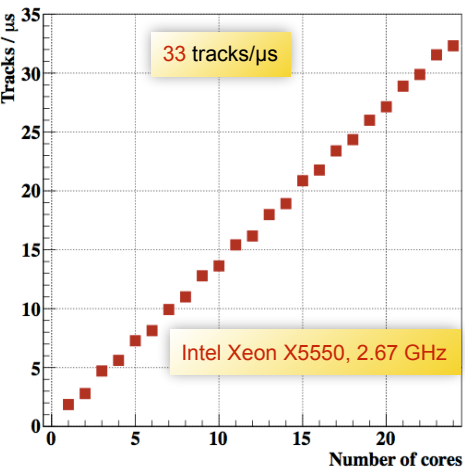


Strong many-core scalability of the Kalman filter library

with I. Kulakov, H. Pabst\* and M. Zyzak (\*Intel)



# Kalman Filter (KF) Track Fit Library



- **Scalability** with respect to the **number of logical cores** in a CPU is one of the most important parameters of the algorithm.
- The scalability on the **Intel Xeon Phi** coprocessor is **similar** to the **CPU**, but running **four threads per core instead of two**.
- In case of the **graphic cards** the set of tasks is divided into **working groups** and **distributed among compute units** (or streaming multiprocessors) and the **load of each compute unit** is of the particular **importance**.

Full portability of the Kalman filter library

# Cellular Automaton (CA) Track Finder

0. Hits (CBM)

1000 Hits

0. Hits

1. Segments

2. Counters

3. Track Candidates

4. Tracks

Detector layers

Hits

Cellular Automaton:

1. Build short track segments.
2. Connect according to the track model, estimate a possible position on a track.
3. Tree structures appear, collect segments into track candidates.
4. Select the best track candidates.

Cellular Automaton:

- local w.r.t. data
- intrinsically parallel
- extremely simple
- very fast

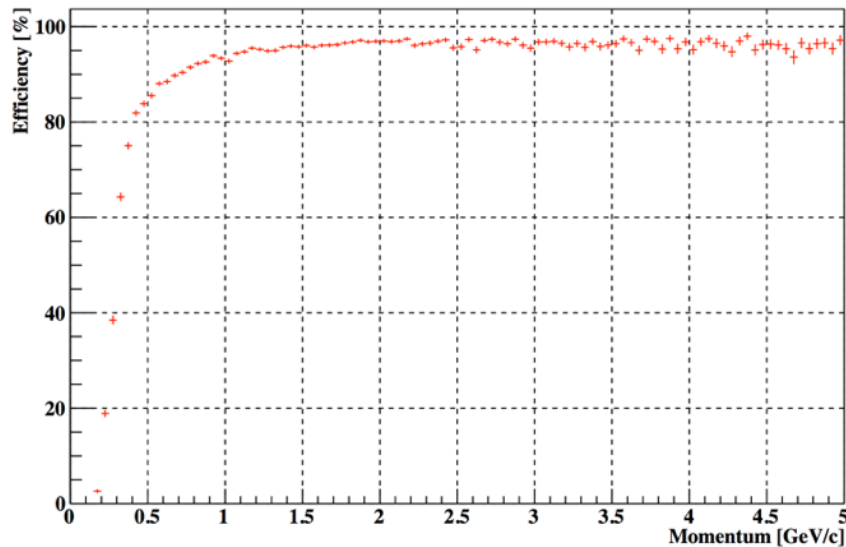
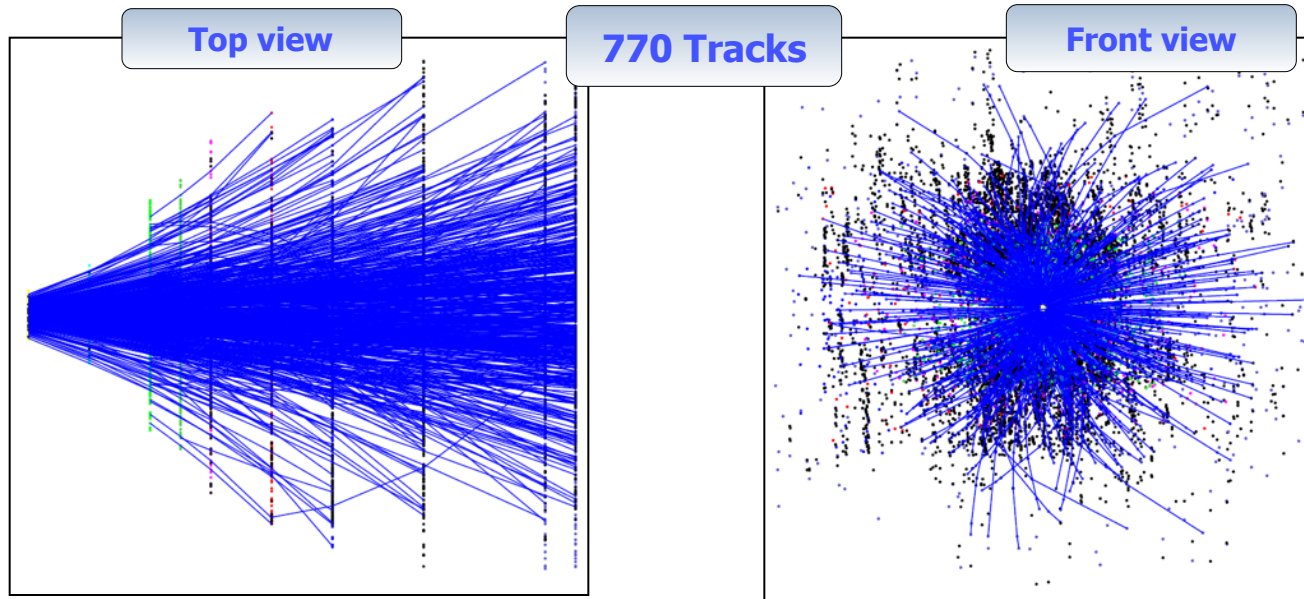
Perfect for many-core CPU/GPU !

4. Tracks (CBM)

1000 Tracks

Useful for complicated event topologies with large combinatorics and for parallel hardware

# CA Track Finder: Efficiency

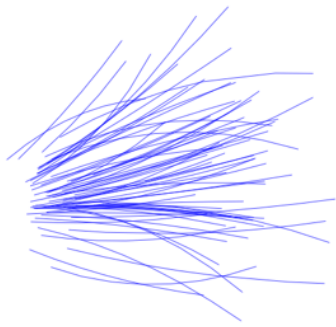


	Efficiency, %	
	mbias	central
Primary high- $p$ tracks	97.1	96.2
Primary low- $p$ tracks	90.4	90.7
Secondary high- $p$ tracks	81.2	81.4
Secondary low- $p$ tracks	51.1	50.6
All tracks	88.5	88.3
Clone level	0.2	0.2
Ghost level	0.7	1.5
Reconstructed tracks/event	120	591
Time/event/core	8.2 ms	57 ms

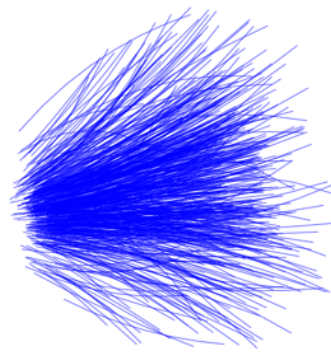
Efficient and stable event reconstruction

# CA Track Finder at High Track Multiplicity

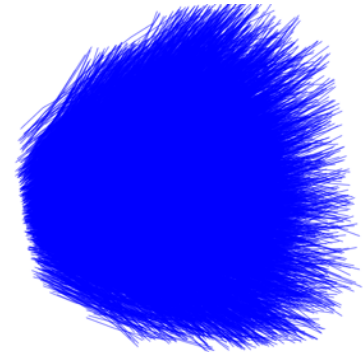
A number of minimum bias events is gathered into a group (super-event), which is then treated by the CA track finder as a single event



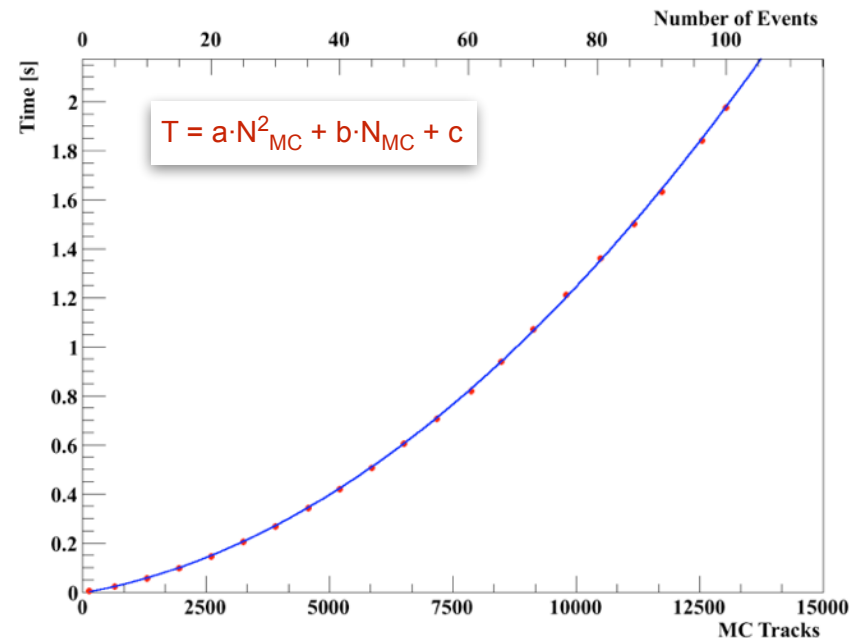
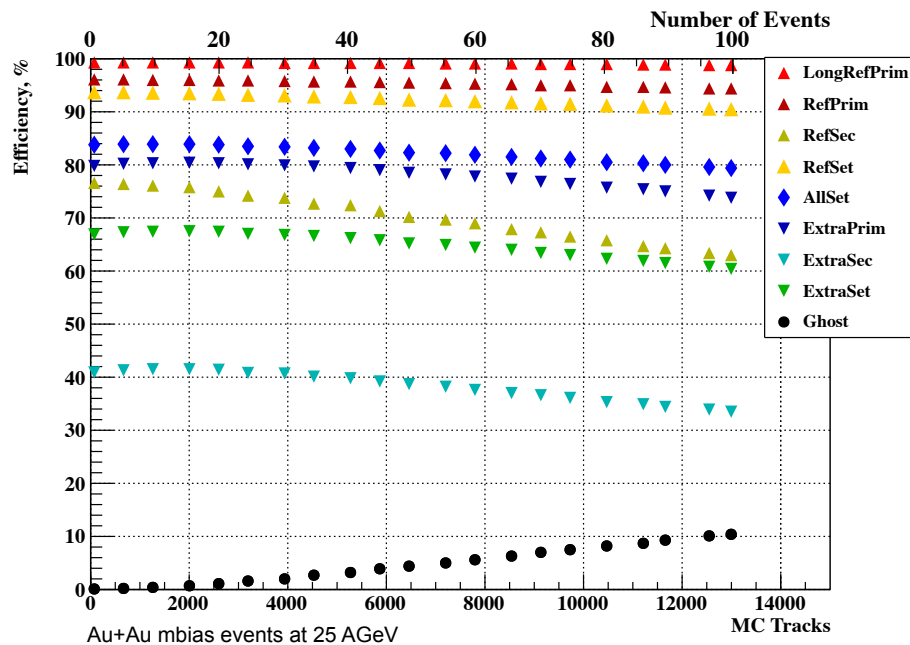
1 mbias event,  $\langle N_{\text{reco}} \rangle = 109$



5 mbias events,  $\langle N_{\text{reco}} \rangle = 572$



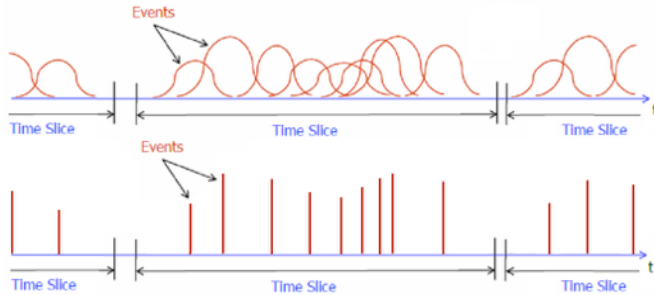
100 mbias events,  $\langle N_{\text{reco}} \rangle = 10340$



Stable reconstruction efficiency and time as a second order polynomial w.r.t. to track multiplicity



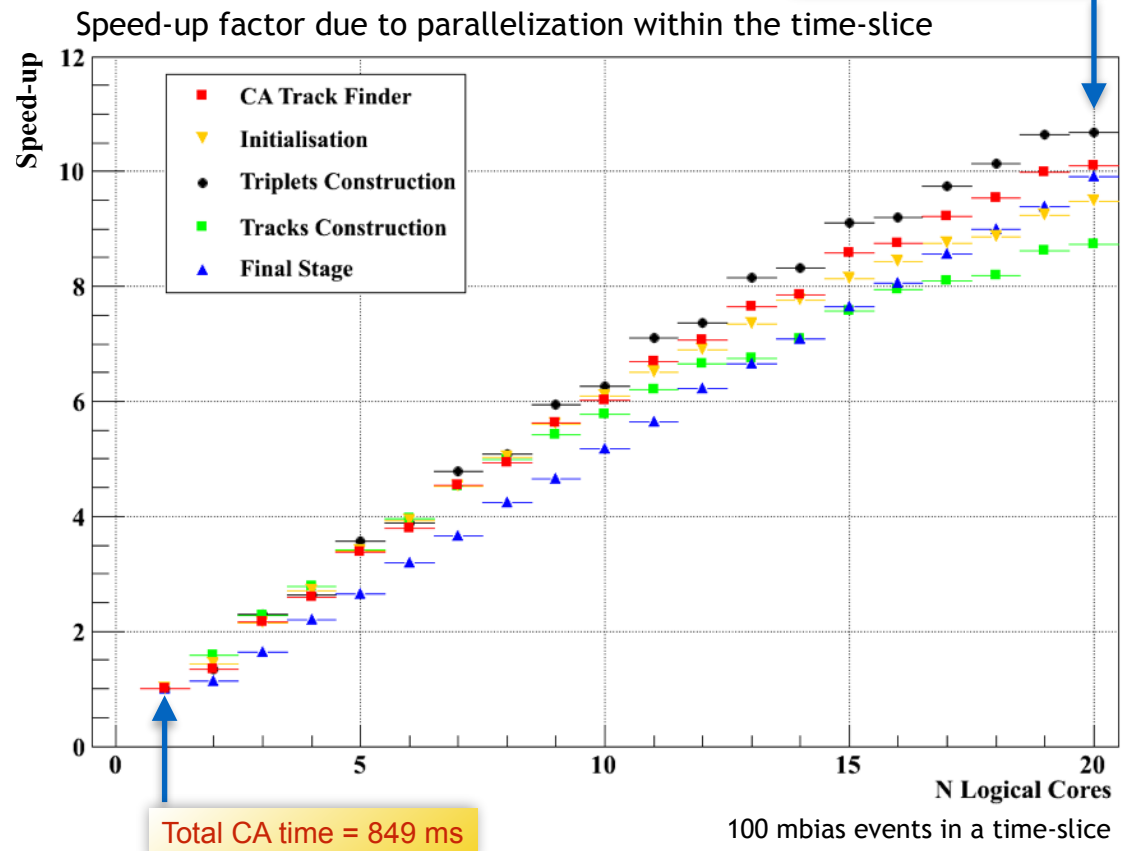
# Time-based (4D) Track Reconstruction with CA Track Finder



- The **beam** in the CBM will have **no bunch structure**, but continuous.
- Measurements in this case will be **4D** ( $x, y, z, t$ ).
- Significant **overlapping of events** in the detector system.
- Reconstruction of **time slices** rather than events is needed.

Stage of the algorithm	% of total execution time
Initialisation	8
Triplets construction	64
Tracks construction	15
Final cleaning	13

Efficiency, %	3D	3+1 D	4D
All tracks	83.8	80.4	83.0
Primary high- $p$	96.1	94.3	92.8
Primary low- $p$	79.8	76.2	83.1
Secondary high- $p$	76.6	65.1	73.2
Secondary low- $p$	40.9	34.9	36.8
Clone level	0.4	2.5	1.7
Ghost level	0.1	8.2	0.3
Time/event/core, ms	8.2	31.5	8.5

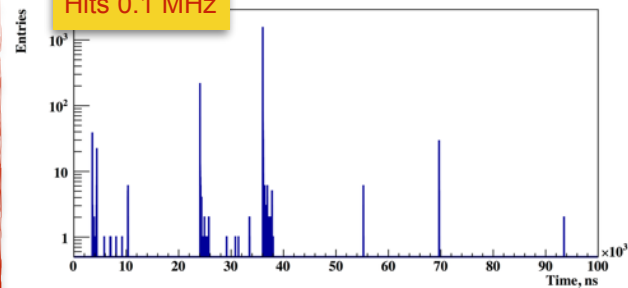


4D event building is scalable with the speed-up factor of 10.1; 3D reconstruction time 8.2 ms/event is recovered in 4D case

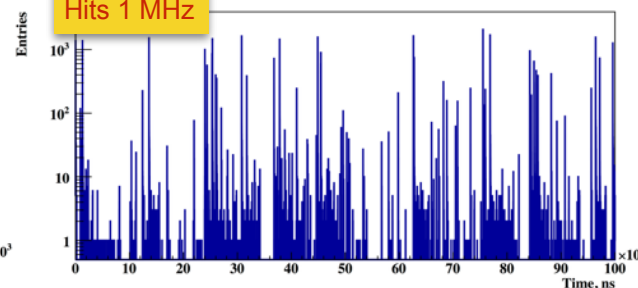
# 4D Event Building at 10 MHz

## Hits at high input rates

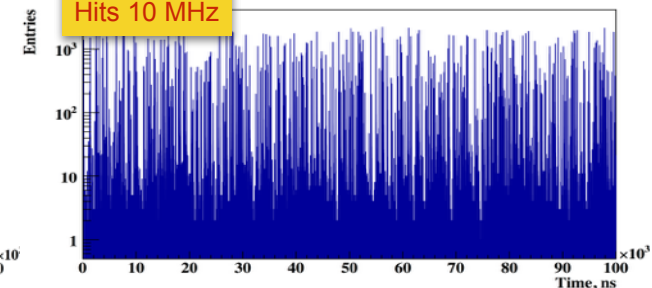
Hits 0.1 MHz



Hits 1 MHz

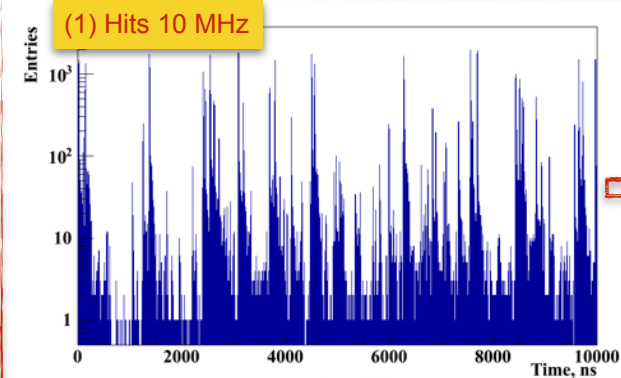


Hits 10 MHz

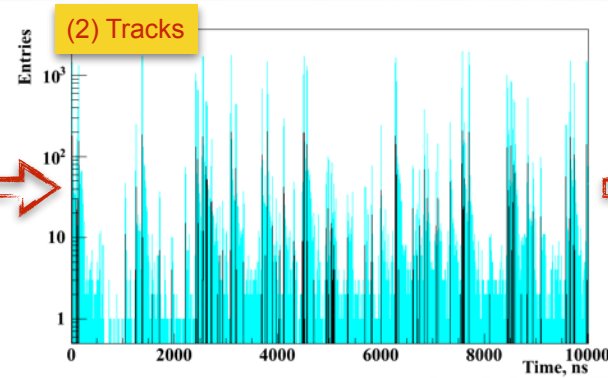


## From hits to tracks to events

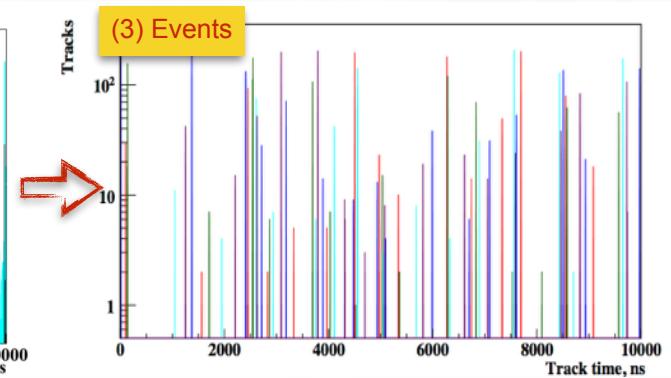
(1) Hits 10 MHz



(2) Tracks



(3) Events



Reconstructed tracks clearly represent groups, which correspond to the original events  
83% of single events, no splitted events, further analysis with TOF information at the vertexing stage

# Summary

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- The Kalman Filter track fit library is vectorized, parallelized and portable to CPU/Phi/GPU architectures.
- The Cellular Automaton track finder is vectorized, parallelized and updated for time-based (4D) track finding in time-slices.
- 4D event building is done after all tracks in the time-slice are found.